

Invariant Subspaces: Controllability and Observability

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Geometric Control Theory for Linear Systems

Block 1: Foundations [10:30 - 12.30]:

- Talk 1: *Motivation and historical perspective*, **G. Marro** [10:30 - 11:00]
- Talk 2: *Invariant subspaces*, **L. Ntogramatzidis** [11:00 - 11:30]
- Talk 3: *Controlled invariance and invariant zeros*, **D. Prattichizzo** [11:30 - 12:00]
- Talk 4: *Conditioned invariance and state observation*, **F. Morbidi** [12:00 - 12:30]

Block 2: Problems and applications [15:30 - 17.30]:

- Talk 5: *Stabilization and self-bounded subspaces*, **L. Ntogramatzidis** [15:30 - 16:00]
- Talk 6: *Disturbance decoupling problems*, **L. Ntogramatzidis** [16:00 - 16:30]
- Talk 7: *LQR and H_2 control problems*, **D. Prattichizzo** [16:30 - 17:00]
- Talk 8: *Spectral factorization and H_2 -model following*, **F. Morbidi** [17:00 - 17:30]

Outline

- **Basic operations on subspaces**
- A -invariant subspaces
- The most important solved problems

State-Space Models

Consider the LTI system Σ ruled by

$$\Sigma : \quad \begin{cases} \dot{x}(t) = A x(t) + B u(t) & x(0) = x_0 \\ y(t) = C x(t) + D u(t) \end{cases}$$

where, for all $t \geq 0$,

- $x(t) \in \mathbb{R}^n$ is the state
- $u(t) \in \mathbb{R}^m$ is the control input
- $y(t) \in \mathbb{R}^p$ is the output.

Subspaces

A subspace \mathcal{X} of \mathbb{R}^n can be represented by a **basis matrix** X , i.e., a matrix whose columns are linearly independent and span \mathcal{X} .

Subspaces

A subspace \mathcal{X} of \mathbb{R}^n can be represented by a **basis matrix** X , i.e., a matrix whose columns are linearly independent and span \mathcal{X} .

Hence, a basis matrix X of \mathcal{X} is such that

$$\begin{aligned}\text{im } X &= \mathcal{X} \\ \text{ker } X &= \{0\}\end{aligned}$$

For computational robustness, we require the columns of X to be an orthonormal basis of \mathcal{X} .

Basic Operations

Matrix $A \in \mathbb{R}^{m \times n}$ is identified with the corresponding linear map between \mathbb{R}^n and \mathbb{R}^m .

$\mathcal{X}, \mathcal{Y}, \mathcal{Z}$: subspaces of \mathbb{R}^n .

\mathcal{H} : subspace of \mathbb{R}^m .

Basic operations:

- *sum*: $\mathcal{X} + \mathcal{Y} = \{x + y \in \mathbb{R}^n \mid x \in \mathcal{X} \text{ and } y \in \mathcal{Y}\}$
- *linear transformation*: $A\mathcal{X} = \{w \in \mathbb{R}^m \mid \exists x \in \mathcal{X} : Ax = w\}$
- *orthogonal complementation*: $\mathcal{X}^\perp = \{y \in \mathbb{R}^n \mid y^\top x = 0 \forall x \in \mathcal{X}\}$
- *intersection*: $\mathcal{X} \cap \mathcal{Y} = \{z \in \mathbb{R}^n \mid z \in \mathcal{X} \text{ and } z \in \mathcal{Y}\}$
- *inverse linear transformation*: $A^{-1}\mathcal{H} = \{x \in \mathbb{R}^n \mid Ax \in \mathcal{H}\}$
(Notice: matrix A needs not be invertible)

Basic Relations

$$\mathcal{X} \cap (\mathcal{Y} + \mathcal{Z}) \supseteq (\mathcal{X} \cap \mathcal{Y}) + (\mathcal{X} \cap \mathcal{Z})$$

$$\mathcal{X} + (\mathcal{Y} \cap \mathcal{Z}) \subseteq (\mathcal{X} + \mathcal{Y}) \cap (\mathcal{X} + \mathcal{Z})$$

$$(\mathcal{X}^\perp)^\perp = \mathcal{X}$$

$$(\mathcal{X} + \mathcal{Y})^\perp = \mathcal{X}^\perp \cap \mathcal{Y}^\perp$$

$$(\mathcal{X} \cap \mathcal{Y})^\perp = \mathcal{X}^\perp + \mathcal{Y}^\perp$$

$$A(\mathcal{X} \cap \mathcal{Y}) \subseteq A\mathcal{X} \cap A\mathcal{Y}$$

$$A(\mathcal{X} + \mathcal{Y}) = A\mathcal{X} + A\mathcal{Y}$$

$$A^{-1}(\mathcal{H}_1 \cap \mathcal{H}_2) = A^{-1}\mathcal{H}_1 \cap A^{-1}\mathcal{H}_2$$

$$A^{-1}(\mathcal{H}_1 + \mathcal{H}_2) \supseteq A^{-1}\mathcal{H}_1 + A^{-1}\mathcal{H}_2$$

$$(A^{-1}\mathcal{H})^\perp = A^T\mathcal{H}^\perp$$

$$A\mathcal{X} \subseteq \mathcal{H} \Leftrightarrow A^T\mathcal{H}^\perp \subseteq \mathcal{X}^\perp$$

Grassman Manifold

For every pair of subspaces $\mathcal{X}_1, \mathcal{X}_2$ of \mathcal{X} ,

- 1 $\mathcal{X}_1 + \mathcal{X}_2$ is the smallest subspace of \mathcal{X} containing both \mathcal{X}_1 and \mathcal{X}_2
- 2 $\mathcal{X}_1 \cap \mathcal{X}_2$ is the largest subspace of \mathcal{X} contained in both \mathcal{X}_1 and \mathcal{X}_2

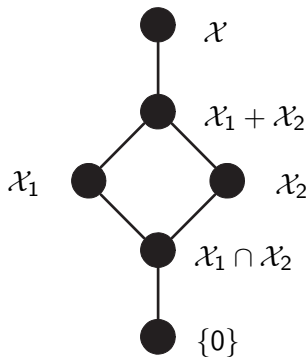


Image in MATLAB[®]

The overall numerical robustness is held up by the routine `ima.m`, based on the QR decomposition.

`>> Q=ima(X, [flag])`: routine `ima.m` performs the orthonormalization of a set of vectors given as the columns of the matrix X and returns them as the columns of matrix Q .

- `flag = 0`: re-ordering of vectors during the orthonormalization process is **not** allowed
- Re-ordering is allowed otherwise

Example:

$$A = \begin{bmatrix} 1 & 1 & 2 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{bmatrix} \rightarrow \text{ima}(A) = \begin{bmatrix} -0.8165 & 0 \\ -0.4082 & -0.7071 \\ 0 & 0 \\ -0.4082 & 0.7071 \end{bmatrix}, \quad \text{ima}(A, 0) = \begin{bmatrix} 0.7071 & -0.4082 \\ 0 & -0.8165 \\ 0 & 0 \\ 0.7071 & 0.4082 \end{bmatrix}$$

Orthogonal Complement in MATLAB®

$$\mathcal{X}^\perp = \{y \in \mathbb{R}^n \mid y^T x = 0 \ \forall x \in \mathcal{X}\}$$

>> `Q=ortco(X)`: routine `ortco.m` computes the orthogonal complement of `im X` as

```
[m,n]=size(ima(X,0));
M=ima([X,eye(n)],0); Q=M(:,n+1:m);
```

Example:

$$A = \begin{bmatrix} 1 & 1 & 2 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{bmatrix} \rightarrow \text{ortco}(A) = \begin{bmatrix} 0.5774 & 0 \\ -0.5774 & 0 \\ 0 & 1 \\ -0.5774 & 0 \end{bmatrix}$$

Sum in MATLAB[®]

$$\mathcal{X} + \mathcal{Y} = \{x + y \in \mathbb{R}^n \mid x \in \mathcal{X} \text{ and } y \in \mathcal{Y}\}$$

>> `Q=sums(X,Y)` routine `sums.m` computes the sum of $\text{im } A$ and $\text{im } B$ as
`Q=ima([X,Y]);`

Example:

$$A = \begin{bmatrix} 1 & 1 & 2 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} 1 \\ 1 \\ 0 \\ 1 \end{bmatrix}, \quad \rightarrow \quad \text{sums}(A, B) = \begin{bmatrix} 0.7071 & 0.4082 & 0.5774 \\ 0 & 0.8165 & -0.5774 \\ 0 & 0 & 0 \\ 0.7071 & -0.4082 & -0.5774 \end{bmatrix}$$

Intersection in MATLAB[®]

$$\mathcal{X} \cap \mathcal{Y} = \{z \in \mathbb{R}^n \mid z \in \mathcal{X} \text{ and } z \in \mathcal{Y}\}$$

We exploit

$$(\mathcal{X} + \mathcal{Y})^\perp = \mathcal{X}^\perp \cap \mathcal{Y}^\perp \quad \implies \quad \mathcal{X} \cap \mathcal{Y} = (\mathcal{X}^\perp + \mathcal{Y}^\perp)^\perp$$

>> `Q=ints(X,Y)` routine `ints.m` computes the intersection of $\text{im } X$ and $\text{im } Y$ as

$$Q = \text{ortco}(\text{sums}(\text{ortco}(X), \text{ortco}(Y)));$$

Example:

$$A = \begin{bmatrix} 1 & 1 & 2 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} 0 & 1 \\ 1 & 0 \\ 0 & 0 \\ -1 & 0 \end{bmatrix}, \quad \rightarrow \quad \text{ints}(A, B) = \begin{bmatrix} 0 \\ -0.7071 \\ 0 \\ 0.7071 \end{bmatrix}$$

Inverse Map in MATLAB[®]

$$A^{-1} \mathcal{H} = \{x \in \mathbb{R}^n \mid Ax \in \mathcal{H}\}$$

>> `Q=invtr(A,H)` routine `invtr.m` computes the subspace $A^{-1} \mathcal{H}$, inverse transform of $\mathcal{H} = \text{im}H$ with respect to the linear map A . It exploits

$$A^{-1} \mathcal{H} = (A^T \mathcal{H}^\perp)^\perp$$

$$Q = \text{ortco}(A' * \text{ortco}(H));$$

Example:

$$A = \begin{bmatrix} 1 & 1 & 2 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{bmatrix}, \quad H = \begin{bmatrix} 4 \\ 3 \\ 0 \\ 1 \end{bmatrix} \quad \rightarrow \quad \text{invtr}(A, H) = \begin{bmatrix} 0.5976 & 0 \\ 0.3586 & 0.8944 \\ -0.7171 & 0.4472 \end{bmatrix}$$

Null-space in MATLAB[®]

$$\ker A = \{x \in \mathbb{R}^n \mid Ax = 0\}$$

>> $Q = \ker(A)$ routine `ker.m` computes the null-space of matrix A through

$$\ker A = (\text{im } A^T)^\perp$$

$$Q = \text{ortco}(A');$$

Example:

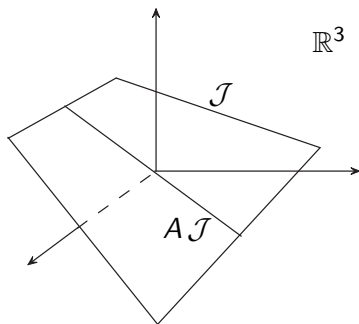
$$A = \begin{bmatrix} 1 & 1 & 2 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{bmatrix} \rightarrow \ker(A) = \begin{bmatrix} -0.5774 \\ -0.5774 \\ 0.5774 \end{bmatrix}$$

Invariant Subspaces

Given $A \in \mathbb{R}^{m \times n}$, a subspace \mathcal{J} of \mathbb{R}^n is an A -invariant if

$$A\mathcal{J} \subseteq \mathcal{J}$$

i.e., $Ax \in \mathcal{J}$ for all $x \in \mathcal{J}$.

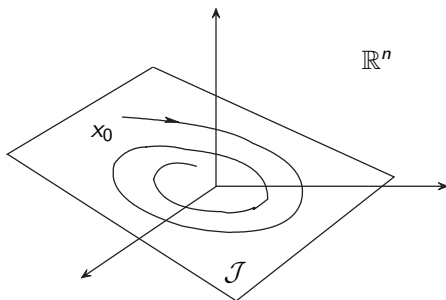


Properties of Invariant Subspaces

Let J be a basis matrix of \mathcal{J} ; the following are equivalent:

- 1 \mathcal{J} is A -invariant;
- 2 \mathcal{J} is a locus of state trajectories of

$$\dot{x}(t) = Ax(t), \quad x(0) = x_0 \in \mathcal{J}$$



Decomposition of Invariant Subspaces

Let J be a basis matrix of the A -invariant \mathcal{J} . Let $T = [J \ T_c]$, with T_c such that T is non-singular. Then

$$A' = T^{-1}AT = \begin{bmatrix} A'_{11} & A'_{12} \\ O & A'_{22} \end{bmatrix}$$

The system written w.r.t. these coordinates is

$$\begin{aligned} \dot{x}_1(t) &= A'_{11} x_1(t) + A'_{12} x_2(t) & x_1(0) &= x_{10} \\ \dot{x}_2(t) &= A'_{22} x_2(t) & x_2(0) &= x_{20} \end{aligned}$$

The MATLAB[®] routine

```
>> [P,Q]=stabi(A,J)
```

computes the matrices $P = A'_{11}$ of $Q = A'_{22}$.

Internal Components

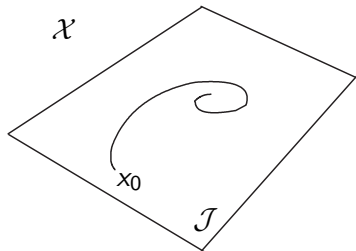
$$\dot{x}_1(t) = A'_{11} x_1(t) + A'_{12} x_2(t) \quad x_1(0) = x_{10}$$

$$\dot{x}_2(t) = A'_{22} x_2(t) \quad x_2(0) = x_{20}$$

If $x_{20} = 0$ ($x(0) \in \mathcal{J}$), then $x_2(t) = 0 \quad \forall t$: the motion on \mathcal{J} is described only by A'_{11} :

$$\dot{x}_1(t) = A'_{11} x_1(t), \quad x_1(0) = x_{10}$$

If A'_{11} is stable, $x(t) \xrightarrow{t \rightarrow \infty} 0$:



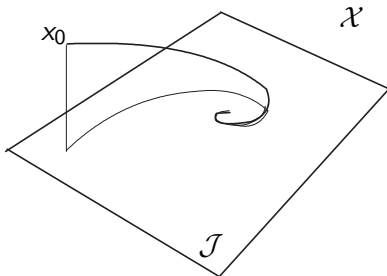
External Components

$$\dot{x}_1(t) = A'_{11} x_1(t) + A'_{12} x_2(t) \quad x_1(0) = x_{10}$$

$$\dot{x}_2(t) = A'_{22} x_2(t) \quad x_2(0) = x_{20}$$

If $x_{20} \neq 0$ ($x(0) \notin \mathcal{J}$), the state trajectory converges to \mathcal{J} if and only if submatrix A'_{22} is stable:

$$\dot{x}_2(t) = A'_{22} x_2(t), \quad x_2(0) = x_{20} \neq 0$$



Complementability

An A -invariant subspace $\mathcal{J} \subseteq \mathcal{X}$ is *complementable* if an A -invariant subspace \mathcal{J}_c exists such that

$$\mathcal{J} \oplus \mathcal{J}_c = \mathcal{X}$$

and \mathcal{J}_c is a *complement* of \mathcal{J} .

Consider again the change of basis $T = [J \ T_c]$. Subspace \mathcal{J} is complementable if and only if the *Sylvester equation*

$$A'_{11} X - X A'_{22} = -A'_{12}$$

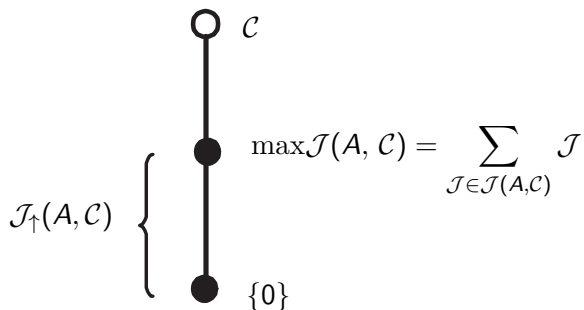
admits a solution. If so, a basis of \mathcal{J}_c is $J_c = JX + T_2$. A sufficient condition is that matrices A'_{11} and A'_{22} have no common eigenvalues.

A-invariants contained in a given subspace

Let \mathcal{C} be a subspace of \mathcal{X} . The set of all the A -invariant subspaces contained in \mathcal{C} is

$$\mathcal{J}(A, \mathcal{C}) = \{ \mathcal{J} \text{ subspace of } \mathcal{X} \mid A\mathcal{J} \subseteq \mathcal{J} \text{ and } \mathcal{J} \subseteq \mathcal{C} \}$$

The set $\mathcal{J}(A, \mathcal{C})$ is closed under addition, hence it admits a maximum:



Computation of $\max \mathcal{J}(A, \mathcal{C})$

Consider the sequence

$$\begin{aligned} \mathcal{Z}_1 &= \mathcal{C} \\ \mathcal{Z}_i &= \mathcal{C} \cap A^{-1} \mathcal{Z}_{i-1} \quad i = 2, 3, \dots \end{aligned}$$

$\max \mathcal{J}(A, \mathcal{C})$ is obtained when the sequence stops (i.e., when $\mathcal{Z}_{i+1} = \mathcal{Z}_i$).

The MATLAB[®] routine

```
>> Q=maxinv(A,C)
```

computes a basis matrix Q of $\max \mathcal{J}(A, \text{im} \mathcal{C})$.

A-invariants containing a given subspace

Let \mathcal{B} be a subspace of \mathcal{X} . The set of all the A -invariant subspaces containing \mathcal{B} is

$$\mathcal{J}(A, \mathcal{B}) = \{ \mathcal{J} \text{ subspace of } \mathcal{X} \mid A\mathcal{J} \subseteq \mathcal{J} \text{ and } \mathcal{J} \supseteq \mathcal{B} \}$$

$\mathcal{J}(A, \mathcal{B})$ is closed under intersection, hence it admits a minimum:

$$\mathcal{J}(A, \mathcal{B}) \left\{ \begin{array}{l} \bullet \quad \mathcal{X} \\ \text{---} \\ \bullet \quad \min \mathcal{J}(A, \mathcal{B}) = \bigcap_{\mathcal{J} \in \mathcal{J}(A, \mathcal{B})} \mathcal{J} \\ \text{---} \\ \circ \quad \mathcal{B} \end{array} \right.$$

Computation of $\min \mathcal{J}(A, \mathcal{B})$

Consider the sequence

$$\begin{aligned} \mathcal{Z}_1 &= \mathcal{B} \\ \mathcal{Z}_i &= \mathcal{B} + A \mathcal{Z}_{i-1} \quad i = 2, 3, \dots \end{aligned}$$

$\min \mathcal{J}(A, \mathcal{B})$ is obtained when the sequence stops (i.e., when $\mathcal{Z}_{i+1} = \mathcal{Z}_i$).

The MATLAB[®] routine

```
>> Q=mininv(A,B)
```

computes a basis matrix Q of $\min \mathcal{J}(A, \text{im}B)$.

Duality

The following dualities hold

$$\begin{aligned}\max \mathcal{J}(A, \mathcal{C}) &= \left(\min \mathcal{J}(A^\top, \mathcal{C}^\perp) \right)^\perp \\ \min \mathcal{J}(A, \mathcal{B}) &= \left(\max \mathcal{J}(A^\top, \mathcal{B}^\perp) \right)^\perp\end{aligned}$$

Controllability

Consider

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t)\end{aligned}$$

The reachable subspace of (A, B) is the set of all the states that can be reached from the origin in any finite time, is the minimum A -invariant containing $\text{im } B$, i.e.,

$$\mathcal{R} = \min \mathcal{J}(A, \text{im } B)$$

- If $\mathcal{R} = \mathcal{X}$, the pair (A, B) is said to be *completely controllable*.
- If $\mathcal{R} \neq \mathcal{X}$ but \mathcal{R} is externally stable, the pair (A, B) is said to be *stabilizable*.

Observability

Consider

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t)\end{aligned}$$

The unobservable subspace of (A, C) is the set of all the initial states that cannot be recognized from the output function in any finite time interval, is the maximum A -invariant contained in $\ker C$, i.e.,

$$Q = \max \mathcal{J}(A, \ker C)$$

- If $Q = \{0\}$, the pair (A, C) is said to be *completely observable*.
- If $Q \neq \{0\}$ but Q is internally stable, the pair (A, C) is said to be *detectable*.